

## Communication Theory (5ETB0) Module 9.2

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## Module 9.2

### Presentation Outline

Part I Motivation: Modern Codes

Part II Capacity Proof

# How do we achieve capacity?

## Modern Codes

- Turbo Codes, invented in 1990-1991, “to good to be true”. Used in 3G and 4G standards. [1]
- Low-density parity check codes, invented in 1960 and re-discovered in 1996. Used in WiFi and DVB.
- Polar Codes, first codes with explicit construction that can be proven to achieve the channel capacity. Used in 5G NR. [2]

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### On the Design of Low-Density Parity-Check Codes within 0.0045 dB of the Shannon Limit

Sae-Young Chung, *Member, IEEE*, G. David Forney, Jr., *Fellow, IEEE*, Thomas J. Richardson, and Rüdiger Urbanke

**Abstract**—We develop improved algorithms to construct good low-density parity-check codes that approach the Shannon limit very closely. For rate 1/2, the best code found has a threshold within 0.0045 dB of the Shannon limit of the binary-input additive white Gaussian noise channel. Simulation results with a somewhat simpler code show that we can achieve within 0.04 dB of the Shannon limit at a bit error rate of  $10^{-6}$  using a block length of  $10^7$ .

**Index Terms**—Density evolution, low-density parity-check codes, Shannon limit, sum-product algorithm.

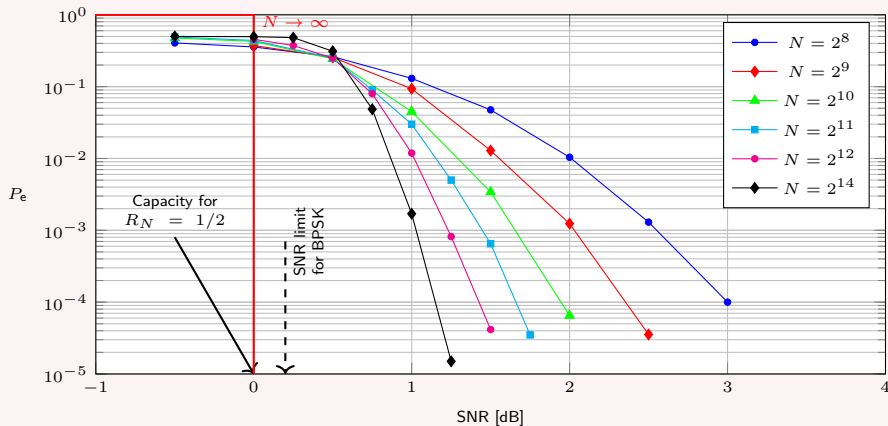
Let  $v$  be a log-likelihood ratio (LLR) message from a degree- $d_v$  variable node to a check node. Under sum-product decoding,  $v$  is equal to the sum of all incoming LLRs; i.e.,

$$v = \sum_{i=0}^{d_v-1} u_i \quad (1)$$

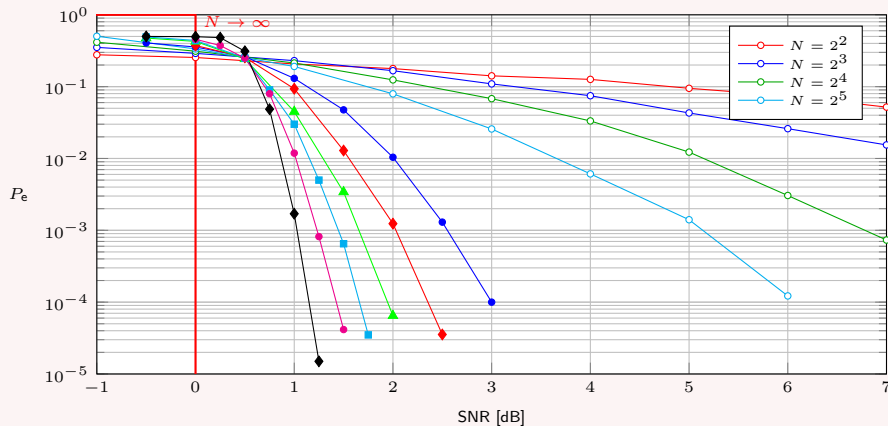
[1] C. Berrou et al., “Near Shannon limit error-correcting coding and decoding: Turbo-codes,” in Proc. IEEE ICC 1993.

[2] E. Arıkan, “Channel Polarization: A Method for Constructing Capacity-Achieving Codes for Symmetric Binary-Input Memoryless Channels,” in IEEE Trans. Inf. Theory, July 2009.

## Example: Polar Codes, $R_N = 1/2$



## Example: Random Codes vs. Polar Codes



# Channel Coding: The Road to Channel Capacity

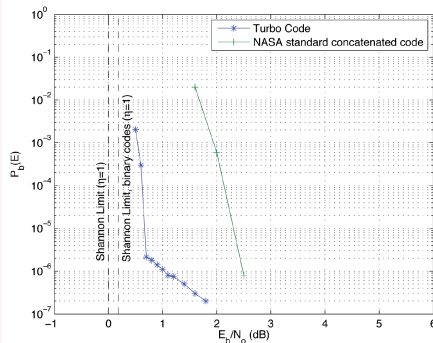
## Turbo Codes vs. NASA Codes



### Channel Coding: The Road to Channel Capacity

*Fifty years of effort and invention have finally produced coding schemes that closely approach Shannon's channel capacity limit on memoryless communication channels.*

By DANIEL J. COSTELLO, JR., *Fellow IEEE*, AND G. DAVID FORNEY, JR., *Life Fellow IEEE*



Excellent review paper [3] where coding history is explained, including NASA codes and the so-called turbo revolution

All this and more in 5LSF0, "Applications of Information Theory" (Q4) and "Information Theory" 5XSE0 (Q3)

[3] Daniel J. Costello, Jr., and G. D. Forney, Jr., "Channel Coding: The Road to Channel Capacity," Proc. of the IEEE, vol. 95, no. 6, June 2007.

## Module 9.2

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Part I Motivation: Modern Codes

Part II Capacity Proof

# Capacity Proof

## Sketch of the Proof

Four key ingredients:

**Part 1:** Sphere Hardening of Gaussian Vectors

**Part 2:** Random Coding Generation

**Part 3:** Everything is Gaussian

**Part 4:** Error Probability



## Part 1: Sphere Hardening of Gaussian Vectors (1/2)

### Gaussian Vectors

Consider a random Gaussian vector  $\underline{G}$  with  $N$  components, each with mean 0 and variance  $\sigma_g^2$  and a normalized version of this vector:  $\underline{G}' = \underline{G}/\sqrt{N}$ . The square norms (lengths) of these vectors ( $\|\underline{G}\|^2$  and  $\|\underline{G}'\|^2$ ) are random variables.

### Sphere Hardening of Gaussian Vectors

The normalized vector  $\underline{G}'$  can be shown to have the following mean and variance:

$$E \left[ \|\underline{G}'\|^2 \right] = \sigma_g^2, \quad \text{var} \left[ \|\underline{G}'\|^2 \right] = \frac{2\sigma_g^4}{N}.$$

Thus, vectors  $\underline{G}'$  are **on the surface of a hypersphere** with radius  $\sigma_g$ . Fluctuations are possible, however for  $N \rightarrow \infty$  these fluctuations disappear.

## Part 1: Sphere Hardening of Gaussian Vectors (2/2)

## Part 2: Random Code Generation

### Generating a Random Code

- Fix the number of signal vectors  $|\mathcal{M}|$  and their number of components (dimensions)  $N$
- Select  $|\mathcal{M}|$  signal vectors  $\underline{s}_1, \underline{s}_2, \dots, \underline{s}_{|\mathcal{M}|}$  *at random*, independently of each other
- Make each vector component is a random sample from a Gaussian density with mean 0 and variance  $E_N$ :
  - By the sphere-hardening argument, *energies of the vectors*  $E[\|\underline{S}\|^2]$  are actually roughly  $NE_N$ .
  - The *expected energy per dimension* is  $E_N$ .
- Consider the *ensemble* of all signal sets that can be chosen in this way

## Part 3: Everything is Gaussian

### Gaussian Inputs, Gaussian Outputs

- The channel output vector is  $\underline{r} = \underline{s}_m + \underline{n}$
- The components of the noise vector  $\underline{n}$  are Gaussian with mean 0 and variance  $N_0/2$
- The sum of two independent Gaussian vectors is also Gaussian, with all components having mean 0 and variance  $E_N + N_0/2$

### Consider Normalized Quantities with $N \rightarrow \infty$

- Output is  $\underline{r}' = \underline{s}'_m + \underline{n}'$ , where  $\underline{s}'_m = \underline{s}_m/\sqrt{N}$  and  $\underline{n}' = \underline{n}/\sqrt{N}$
- Normalized vector  $\underline{s}'_m$  is on the surface of a hypersphere with radius  $\sqrt{E_N}$ , i.e.,  $\|\underline{s}'_m\|^2 = E_N$
- Normalized noise vector  $\underline{n}'$  is on the surface of a hypersphere with radius  $\sqrt{N_0/2}$ , i.e.,  $\|\underline{n}'\|^2 = N_0/2$
- Normalized received vector  $\underline{r}'$  is on the surface of a hypersphere with radius  $\sqrt{E_N + N_0/2}$ , i.e.,  $\|\underline{r}'\|^2 = E_N + N_0/2$

## Part 4: Error Probability

### Average Error Probability

- We are interested in  $P_e^{\text{av}}$ , i.e., the error probability  $P_e$  *averaged over the ensemble of signal sets*, i.e.,

$$P_e^{\text{av}} = \int_{\mathbb{R}^{N \cdot |\mathcal{M}|}} p(\underline{s}_1, \underline{s}_2, \dots, \underline{s}_{|\mathcal{M}|}) P_e(\underline{s}_1, \underline{s}_2, \dots, \underline{s}_{|\mathcal{M}|}) d\underline{s}_1 d\underline{s}_2 \dots d\underline{s}_{|\mathcal{M}|}$$

- Once we know  $P_e^{\text{av}}$  we claim that there exists *at least one signal set*  $\{\underline{s}_1, \underline{s}_2, \dots, \underline{s}_{|\mathcal{M}|}\}$  with error probability  $P_e(\underline{s}_1, \underline{s}_2, \dots, \underline{s}_{|\mathcal{M}|}) \leq P_e^{\text{av}}$
- It can then be shown that if  $R_N = C_N - \delta$

$$\lim_{N \rightarrow \infty} P_e^{\text{av}} \leq \lim_{N \rightarrow \infty} 2^{-\delta N} \sqrt{\frac{E_N + N_0/2}{N_0/2}} = 0$$

## Summary Module 9.2

### Take Home Messages

- Four key ingredients in the capacity proof
- Codes with Gaussian inputs achieve capacity
- Modern codes exist that approach capacity
- Is channel capacity important? Yes because...
  - It is a fundamental limit with beautiful and simple derivations
  - We want to approach such limits
  - Signal shaping

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